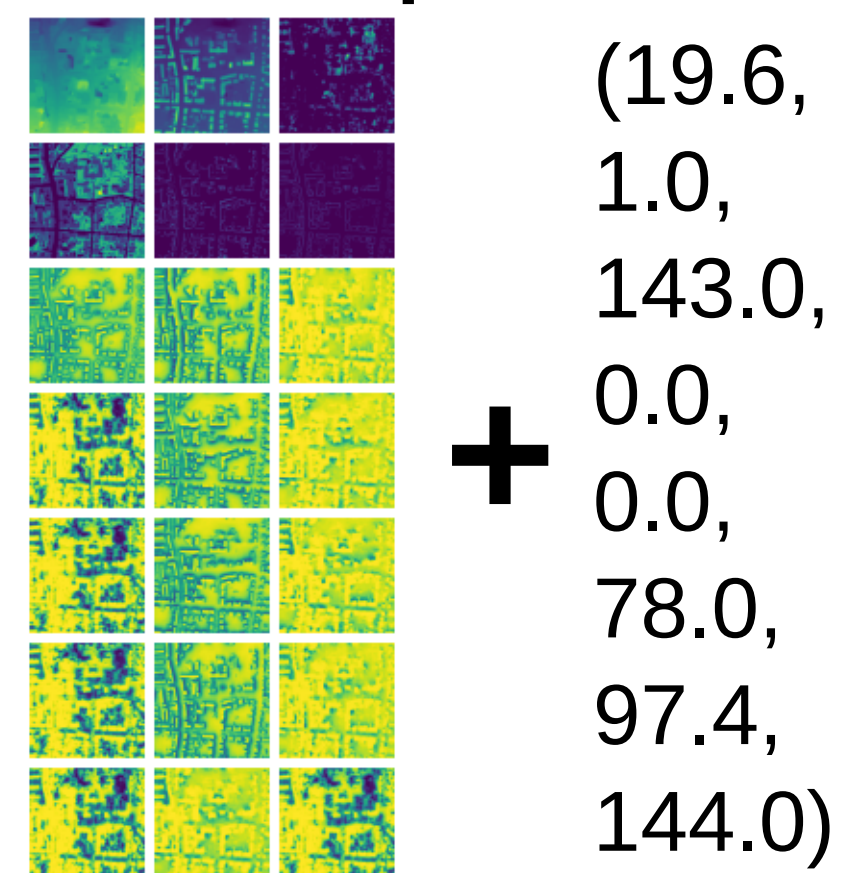


Motivation

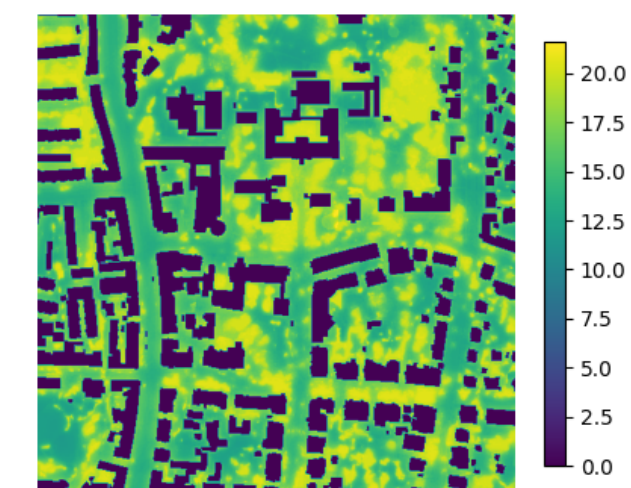
- Physical simulations are currently the tool of choice

Physical inputs



Physical simulation

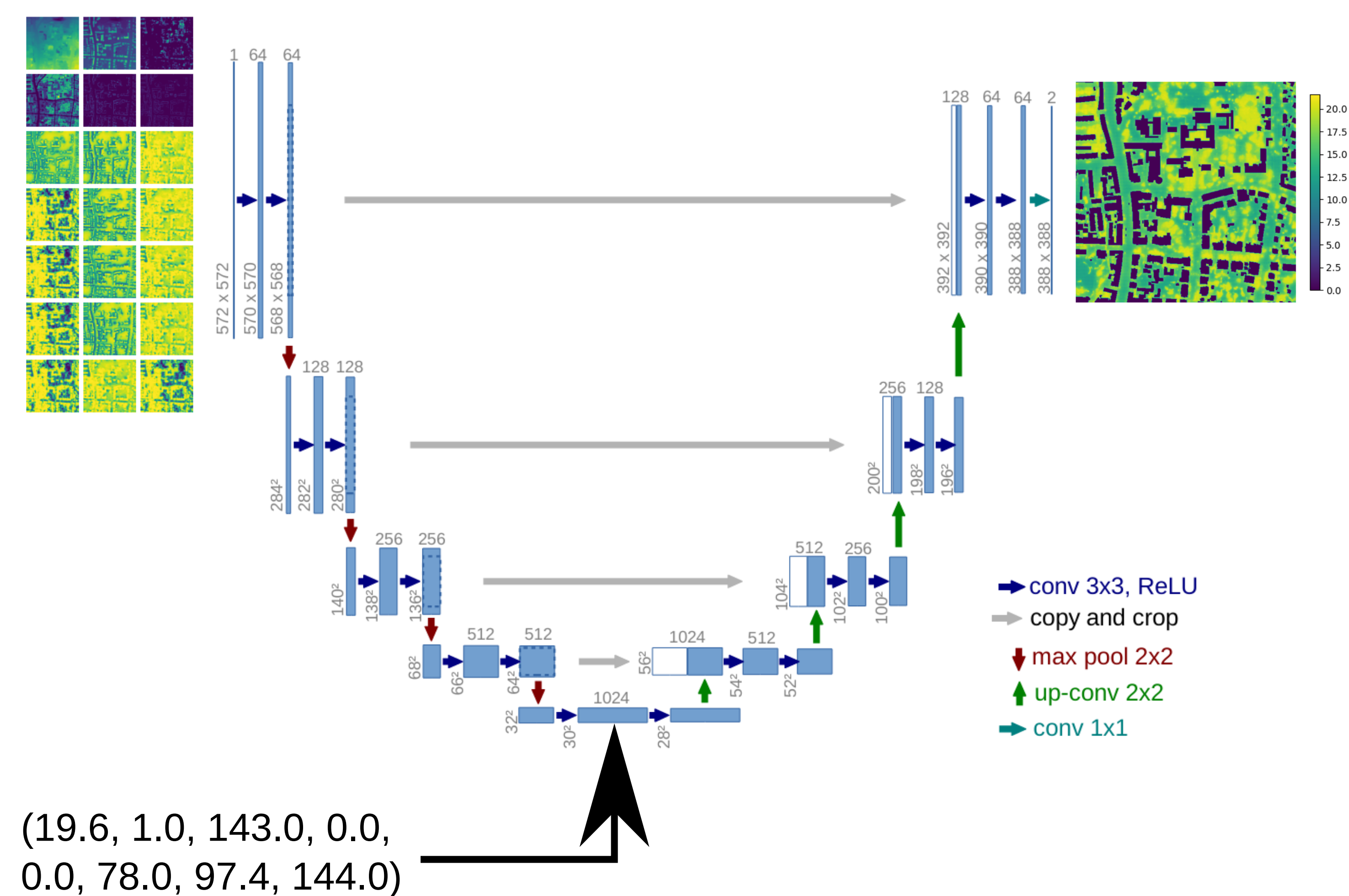
Physical output(s)



- But they are computationally intensive
- Estimating counterfactuals is virtually impossible

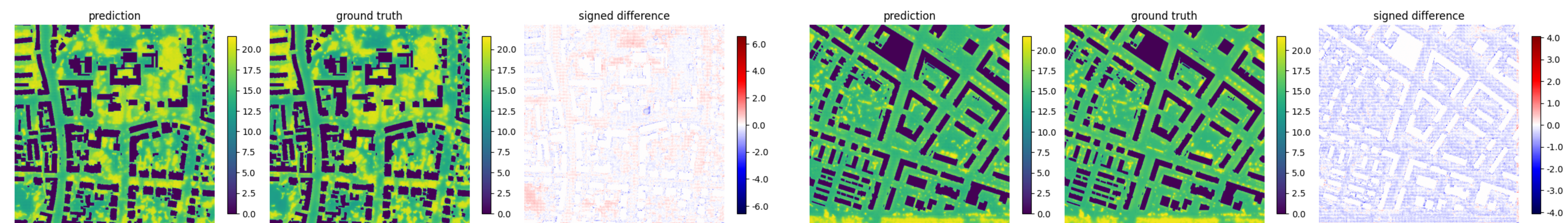
Method

- Deep neural networks achieve strong results
- They learn to approximate a function from data
- We use an encoder-decoder network (U-Net)
- We train the network with simulated data

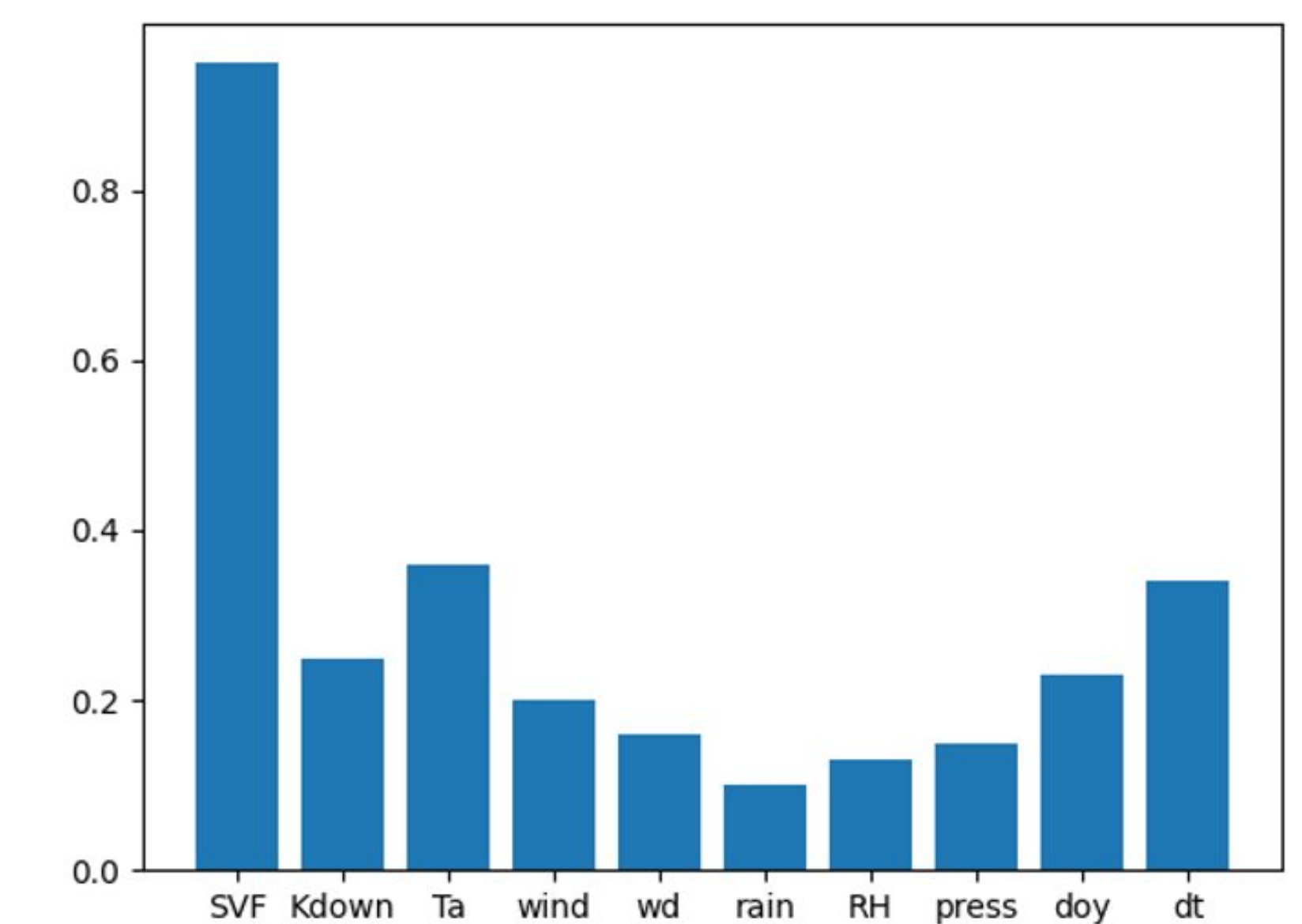
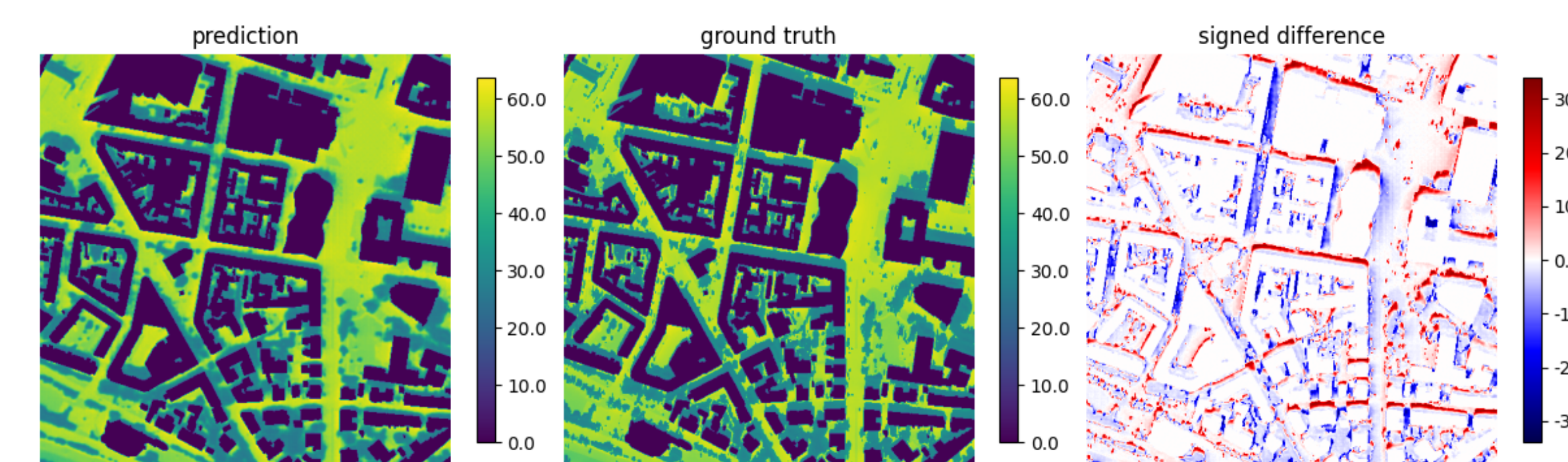


Results

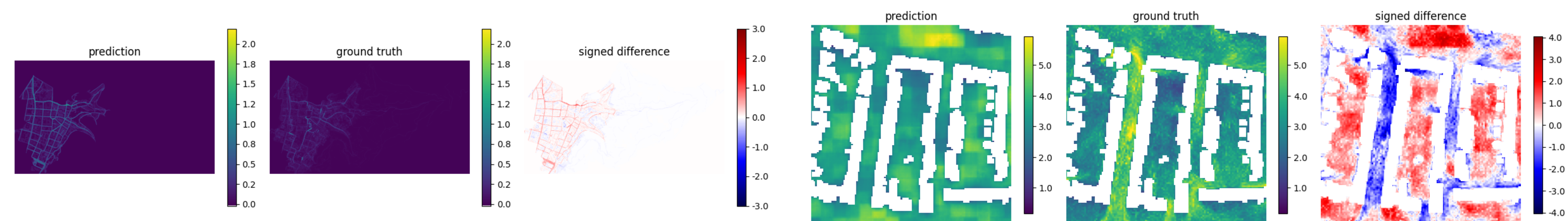
- Heat module yields good results



- But the position of the sun is challenging



- Water and wind modules do not work well yet



Conclusion & future work

- We can achieve good results for the heat module
- But water and wind module are currently a challenge
- Next steps:
 - Improving the function approximation by searching for a neural architecture
 - Incorporating uncertainty in inputs to account for the uncertainty of climate models
 - Estimation of counterfactuals to support urban planning